

Stat 414 – Day 11 Random Slopes cont., Centering (4.6, 5.2)

Last Time: Random slopes (Ch. 5)

- $Y_{ij} = \beta_{00} + u_{0j} + (\beta_{10} + u_{1j})x_{ij} + \epsilon_{ij}$
 - Effect of x_{ij} differs among Level 2 groups (e.g., schools)
 - Adds variance component and covariance between slopes and intercepts
- $Var(Y_{ij}) = \tau_0^2 + 2\tau_{01}x_{ij} + \tau_1^2x_{ij}^2 + \sigma^2$
 - Results impacted by choice of scaling (origin) of x variables
 - Smallest when $x = -\tau_{01}/\tau_1^2$ (correlation vs. covariance)
 - Expect 95% of slopes within $\beta_{10} \pm 2\tau_1$
- Cov of two individuals in same group: $\tau_0^2 + \tau_{01}(x_{aj} + x_{bj}) + \tau_1^2(x_{aj}x_{bj})$
 - No simple ICC (depends on x)
- Explaining level 2 variance
 - Consider intercepts, slopes of "outcomes"

$$Corr = \frac{Cov(X, Y)}{SD(X)SD(Y)}$$

Keep in mind: With random slopes,

$$Corr(Y_{aj}, Y_{bj}) = \frac{Cov(Y_{aj}, Y_{bj}) / SD(Y_{aj})SD(Y_{bj})}{\sqrt{(\tau_0^2 + 2\tau_{01}x_{aj} + \tau_1^2x_{aj}^2 + \sigma^2)(\tau_0^2 + 2\tau_{01}x_{bj} + \tau_1^2x_{bj}^2 + \sigma^2)}}$$

$$\hat{\tau}_{01} = -.86 \times .5296 \times .1386 = -.064$$

For Netherlands data:

<table border="1" style="width: 100%; border-collapse: collapse;"> <thead> <tr> <th colspan="5">Random effects:</th> </tr> <tr> <th>Groups</th> <th>Name</th> <th>Variance</th> <th>Std.Dev.</th> <th>Corr</th> </tr> </thead> <tbody> <tr> <td rowspan="2">school</td> <td>(Intercept)</td> <td>0.28050</td> <td>0.5296</td> <td></td> </tr> <tr> <td>gevocab</td> <td>0.01922</td> <td>0.1386</td> <td>-0.86</td> </tr> <tr> <td>Residual</td> <td></td> <td>3.66613</td> <td>1.9147</td> <td></td> </tr> </tbody> </table> <p>getVarCov(model1b, type = "marginal") cov2cor(cov\$'767') cov2cor(cov[[1]])</p>	Random effects:					Groups	Name	Variance	Std.Dev.	Corr	school	(Intercept)	0.28050	0.5296		gevocab	0.01922	0.1386	-0.86	Residual		3.66613	1.9147		$x_a = 1.7 \quad x_b = 3.1$ $.2805 + (-.064)(4.8) + .01922(1.7)(3.1) = .075$ <hr style="border: 1px solid blue;"/> $\sqrt{.2805 + 2(-.064)(1.7) + (.01922)(1.7^2) + .3666}$ $\times \sqrt{.2805 + 2(-.064)(3.1) + (.01922)(3.1^2) + .3666}$ $= .021$ <p style="text-align: center;">$x_a = 1.7 \quad x_b = 9.3 \Rightarrow -.027$</p>
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Not a simple intraclass correlation coefficient with random slopes.

Example 1: Day 10 Example 2 cont. (Imer)

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$$c \text{ gevocab} = \text{gevocab} - \text{mean}(\text{gevocab})$$

Definition: Grand mean centering is subtracting off the overall mean, $y_{ij} - \bar{y}$. Benefits include interpretation of intercepts, reduce collinearity in "product terms." Can also be especially helpful with strong slope/intercept covariance. "When working with random coefficient models, it is best that **all** independent variables be centered.

Group mean centering is subtracting off the group mean, $y_{ij} - \bar{y}_j$. This is not equivalent and impacts interpretation, but is less common in practice.

(a) Can we include an interaction between age and gevocab in the two cases? How do we interpret this interaction?

(b) How does centering *gevocab* impact our earlier analysis?

(c) What if don't want the random slopes to be correlated?
 (age | school) + (gevocab | school)

Example 2: Recall the data predicting *PctBush* with counties clustered within states. We want to consider the percentage of families in the county in a traditional nuclear form as predictor.

(a) What is the overall mean *PctFamily*? 19.02 \bar{y}

(b) What are the means of *PctFamily* for the first four states?

AL 17.39 AR 17.801 AZ 18.247 CA 21.028 \bar{y}_j

(c) What would *PctFamily* be if grand-mean centered? Group-mean centered?

state_po	county	PctBush	Income	PctFamily	Grand mean centered	Group mean centered
AL	Autauga	75.67352	48.458	25.3	25.3 - 19.02 = 6.28	25.3 - 17.39 = 7.91
AR	Boone	66.27191	34.974	19.6	19.6 - 19.02 = 0.58	19.6 - 17.801 = 1.799
CA	Orange	59.80009	64.611	26.1	26.1 - 19.02 = 7.08	26.1 - 21.028 = 5.072

(d) In a simple model with a fixed slope for *PctFamily*, how do the slope and intercept compare depending on which version you use? Why?

	Intercept	Slope
Raw data (no centering)	40.21	.397
Grand mean	57.29	.397
Group mean	57.01	.398

} same slope but different intercept
 } similar to grand mean centering

(e) In a model with a random slope for *PctFamily*, how do the slope and intercept and slope/intercept correlation values compare for the different versions? What about the prediction for Autauga?

	Intercept	Slope	Corr	\hat{y}
Raw data	39.77	.938	-.34	82.97
Grand mean	57.64	.938	0.10	82.97
Group mean	56.97	.924	-.04	82.88 ← not equivalent

(f) Which models are equivalent in the predicted values? Why?
 raw data (no centering) & grand mean (shifting)
 group mean centering creates a different variable

In considering Level 2 variables to add to the model, an “especially important type” of “contextual variable” is the group mean.

(g) Include pctfamilyavg in the model, along with the interaction with PctFam. *high VIF*

(h) Include StatePctFam, grand-mean centered, in the model and compare interpretations depending on how the Level 1 PctFam variable is centered.

	Intercept	PctFam	StatePctFam	
Raw data	17.4	0.89	1.22	
Both Grand mean	57.5	0.89	1.22	change in PctFam for 2 counties w/ same PctFam but State PctFam differs by 1 <i>y, x, x̄</i>
L1: Group mean/ L2: Grand mean	57.5	0.89*	2.11	Change in mean PctFam when mean State Pct Family differs by 1 <i>y, x̄</i>

Note: Testing the group level coefficient (coefficient of the group mean) is referred to as the Hausman specification test in econometrics.

(i) What does a positive coefficient for the group mean tell you?

w/ grand mean: coefficient β_2 represents additional increase in response w/ higher β_1 at Level 2

(j) What do you notice about the sum of the slope coefficients in row 2 and the slope coefficients in row 3?

0.89 = .89 + 1.22
slope at Level 2 (between group) → within group additional contribution

(k) How do these results compare to the “within group” regression and “between group” regression?

2.45 "effect" w/ one-unit increase in student IQ w/ school mean IQ "fixed"
1.28 is increase in slope for school level effect

(l) How do you summarize and evaluate the significance of the “contextual effect”?
the contextual effect here is the 1.28

language

Example 3: Reconsider the Netherlands data on language score based on verbal IQ from the text (e.g., Section 4.6)

(a) Fit the model with both IQ and mean IQ. What is the slope of the “within group” regression?

2.45

(b) What is the additional contribution to langPOST from the class mean verbal IQ? Is this statistically significant?

1.28

*t = 4.906
 ⇒ statistically significant*

(c) What if we put both grand-mean centered IQ and grand-mean centered mean IQ in the model; how do we interpret the coefficient of mean IQ in this model?

1.28 additional contribution

sometimes called the "contextual effect" - it's what the school contributes to the prediction above and beyond the individual student prediction (two students with same IQ but below to schools that are 1 apart on school average IQ)

(d) How does the coefficient of mean IQ change when we use the group-mean centered verbal IQ?

$$3.74 = 1.28 + 2.45$$

with group mean centering, slope of mean IQ is 3.75. You might prefer this information if your focus is on relationships at the group level (e.g., funding decisions); effect of increase in school average on the response average (see question e)

(e) How does this coefficient compare to the slope of mean IQ in the aggregated data set?

Same idea, similar

If regress 211 language means (one for each school) on 211 IQ means (one for each school), get a slope of 3.78

(f) Interpret the slope/intercept correlation in the multilevel model. (first make IQ have random

random slopes for IQ centered slopes)
as intercepts increase, slopes decrease

(g) What about including SES in the model? (see day 12)